

Information-theoretic inflectional classification

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QUANTITATIVE TYPOLOGY OF INFLECTIONAL CLASSIFICATION

- ▶ Concept of Inflection Classes widely used to analyse inflectional systems

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 - ▶ Large datasets
 - ▶ Reproducible classifications
 - ▶ Commensurable in several languages
 - ▶ Basis for theoretical and typological comparisons.

Groups of lexemes that inflect alike.

	INF	PRES.3.SG	PRES.3.PL	PP
TENIR 'hold'	təniʁ	tjɛ	tjɛn	təny
FINIR 'finish'	finiʁ	fini	finis	fini
HAÏR 'hate'	aïʁ	ɛ	ais	ai
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WHAT WE NEED TO INFER IC FROM PARADIGMATIC DATA.

1. What form should an IC system take ?
2. What generalisations should we infer from the data ?
3. How do we measure which lexemes inflect alike ?
4. How do we find the best classes among all possible ones ?

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INFLECTION CLASSES: COHESIVE OR DISTINCTIVE ?

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 - ▶ **Perfectly cohesive** : *numerous small, similar classes*
 - ▶ **Perfectly distinctive** : *fewer large classes with exceptions*

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INFLECTION CLASSES: COHESIVE OR DISTINCTIVE ?

- ▶ Dressler and Thornton's terminology (1996):
- ▶ **Micro-classes**
 - ▶ *Internally homogeneous : numerous small, similar classes*
- ▶ **Macro-classes**
 - ▶ *Externally heterogeneous : fewer large classes with exceptions*

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 - ▶ *Internally homogeneous : numerous small, similar classes*
- ▶ **Macro-classes**
 - ▶ *Externally heterogeneous : fewer large classes with exceptions*
- ▶ Combined in a hierarchy. (Corbett and Fraser, 1993; Dressler and Thornton, 1996; Brown and Evans, 2012)

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TWO STRATEGIES FOR ABSTRACTING GENERALISATIONS.

- ▶ Stem and exponents /bæg/ + $\left\{ \begin{array}{l} /z/(PL) \\ \emptyset (SG) \end{array} \right\}$
 - ▶ *Captures differences between cells under the assumption of a constant stem.*

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 - ▶ *Captures differences between cells under the assumption of a constant stem.*
- ▶ Binary alternation patterns $/bæg/ (SG) \rightleftharpoons /bægz/ (PL)$
 $/X/ (SG) \rightleftharpoons /Xz/ (PL)$
 - ▶ *Captures the **implicative relation** between each pair of cells.*

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- ▶ Binary alternation patterns $/bæg/ (SG) \rightleftharpoons /bægz/ (PL)$
 $/X/ (SG) \rightleftharpoons /Xz/ (PL)$
 - ▶ *Captures the implicative relation between each pair of cells.*
- ▶ Crucial difference:
 - ▶ global segmentation over the whole paradigm.
 - ▶ local segmentation over pairs of forms.

SEGMENTATION STRATEGIES

- ▶ **Global** : On the basis of a whole paradigm.
- ▶ **Local** : On each pair of cells.

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HAÏR ‘hate’	aíꝝ	ε	ais	ai
PELER ‘peel’	X ₁ əX ₂ e	X ₁ ɛX ₂	X ₁ ɛX ₂	X ₁ əX ₂ e
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FINIR ‘finish’	Xiꝝ ⇌ X	Yiꝝ ⇌ Ys	Ziꝝ ⇌ Z	
HAÏR ‘hate’	aiꝝ ⇌ ε	Yiꝝ ⇌ Ys	Ziꝝ ⇌ Z	...
PELER ‘peel’	X ₁ əX ₂ e ⇌ X ₁ εX ₂	Y ₁ əY ₂ e ⇌ Y ₁ εY ₂	Z ⇌ Z	
LAVER ‘wash’	Xe ⇌ X	Ye ⇌ Y	Z ⇌ Z	
TASSER ‘press’	Xe ⇌ X	Ye ⇌ Y	Z ⇌ Z	

A CLUSTERING PROBLEM

- ▶ This leads us to a **clustering problem** :
 - ▶ Evaluate fitness of clusters
 - ▶ Explore the search space

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DESCRIPTION LENGTH

- ▶ **Minimum description length** (Rissanen, 1984) : Choose the model allowing for the shortest description of the data.
- ▶ A partition of the set of lexemes is better than another one if it leads to a more economical description of the system. (Sagot and Walther, 2011; Walther, 2013)

$$DL(description) =$$

$$\text{length}(description) \times - \sum_{x \in \text{symbols}(description)} P(x) \log_2(P(x))$$

DESCRIPTION LENGTH

- ▶ **Microclasses:** number of bits necessary to encode the mapping between each lexeme and its microclass.

$$M = - \sum_{m \in Microclasses} |m| \times \log_2\left(\frac{|m|}{|lexemes|}\right)$$

DESCRIPTION LENGTH

- ▶ **Microclasses:** number of bits necessary to encode the mapping between each lexeme and its microclass.
- ▶ **Lexicon:** number of bits necessary to encode the mapping between each microclass and its clusters.

$$L = - \sum_{c \in \text{clusters}} |c| \times \log_2 \left(\frac{|c|}{|\text{microclasses}|} \right)$$

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- ▶ **Microclasses:** number of bits necessary to encode the mapping between each lexeme and its microclass.
- ▶ **Lexicon:** number of bits necessary to encode the mapping between each microclass and its clusters.
- ▶ **Grammar:** number of bits necessary to define the list of patterns available for each cluster and each pair of cells.

$$G = - \sum_{c \in \text{clusters}} \sum_{A, B \in \text{pairs(cells)}} |patterns_c^{A,B}| \times \log_2 \left(\frac{|patterns_c^{A,B}|}{|patterns^{A,B}|} \right)$$

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- ▶ **Residual ambiguity:** number of bits necessary to encode the mapping of each microclass to its pattern given the lexicon and the grammar

$$R = - \sum_{c \in \text{clusters}} \sum_{A, B \in \text{pairs(cells)}} \sum_{t \in \text{patterns}_c^{A, B}} |t| \times \log_2\left(\frac{|t|}{|c|}\right)$$

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$$DL = M + L + G + R$$

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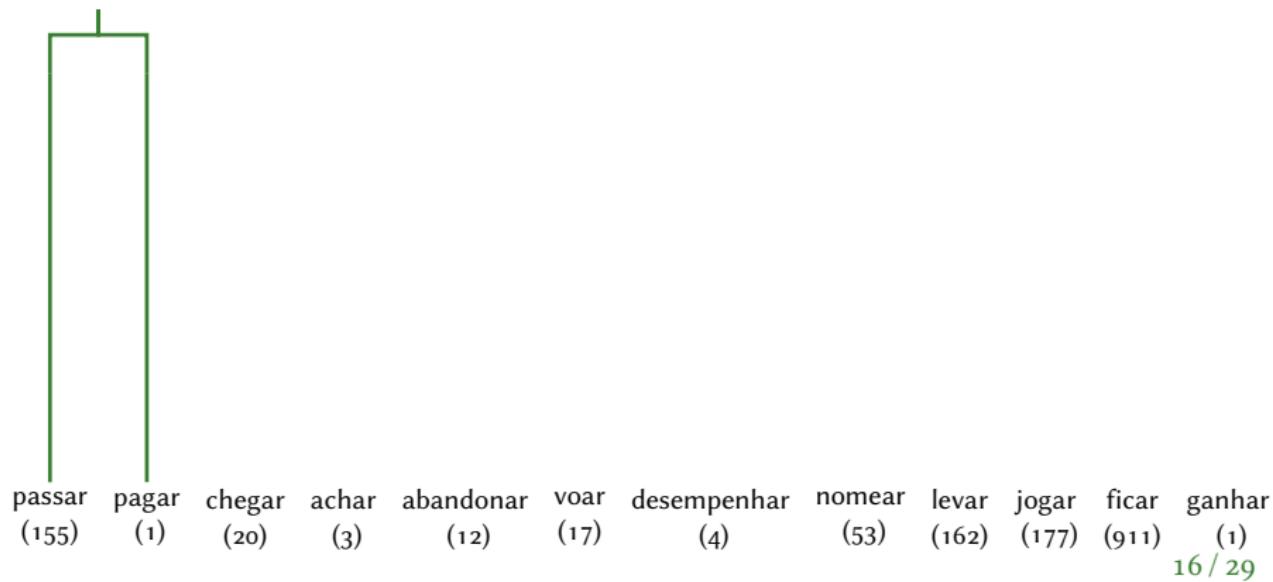
CLUSTERING ALGORITHM

- (a) Merge lexemes with identical patterns (micro-classes).

passar	pagar	chegar	achar	abandonar	voar	desempenhar	nomear	levar	jogar	ficar	ganhar
(155)	(1)	(20)	(3)	(12)	(17)	(4)	(53)	(162)	(177)	(911)	(1)

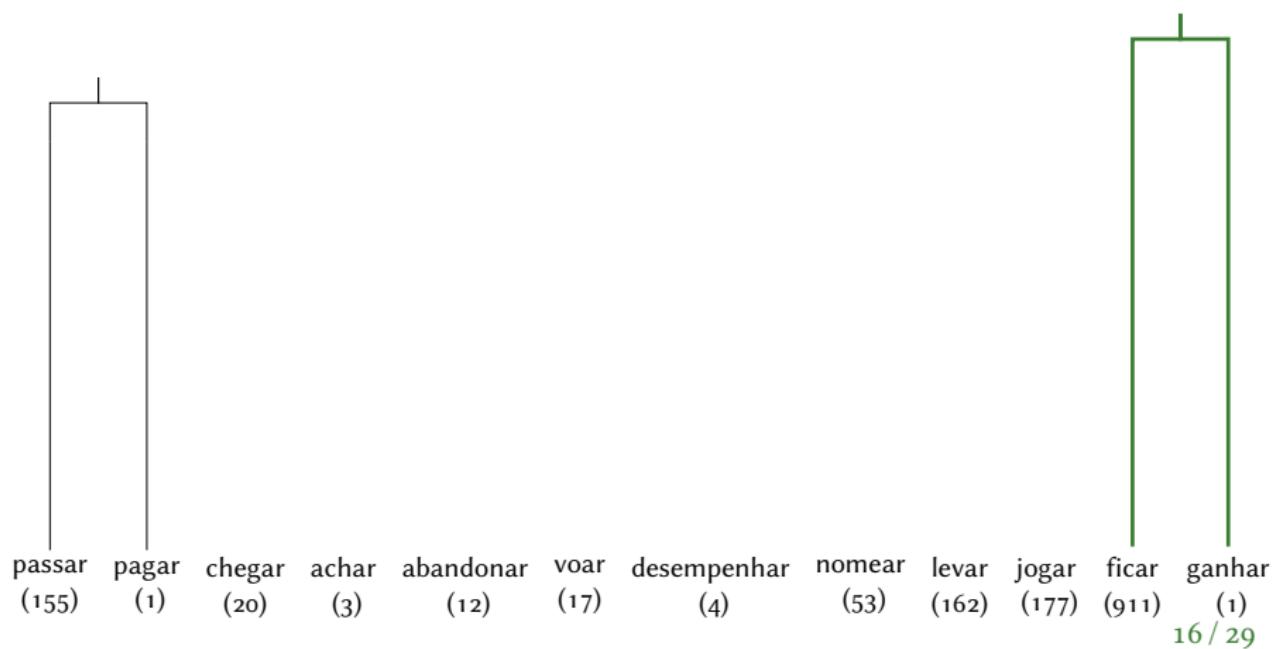
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- (b) Merge the pair optimising descriptive economy (lowest DL).



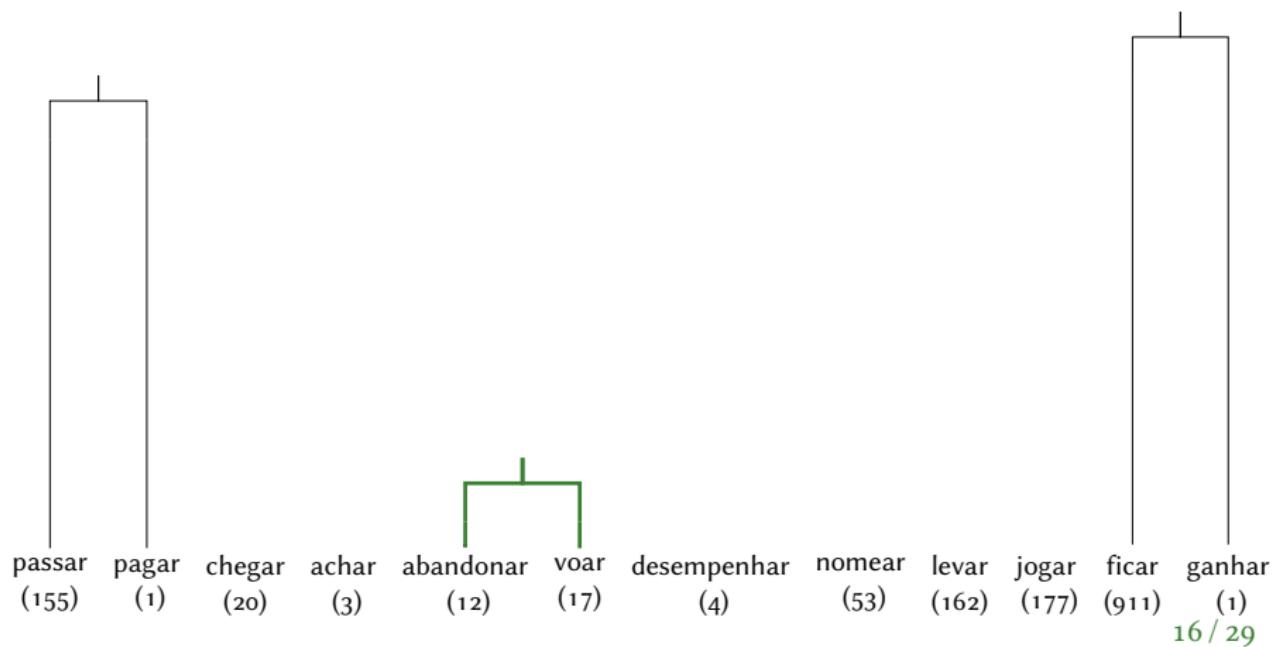
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- (a) Merge lexemes with identical patterns (micro-classes).
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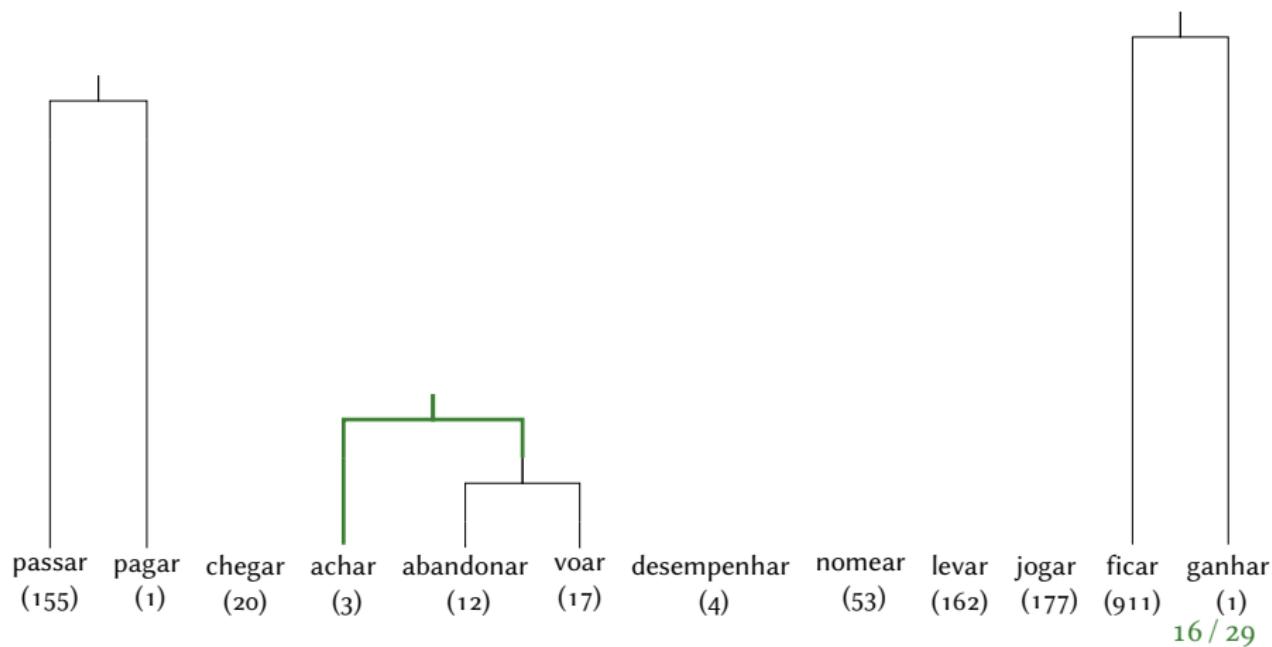
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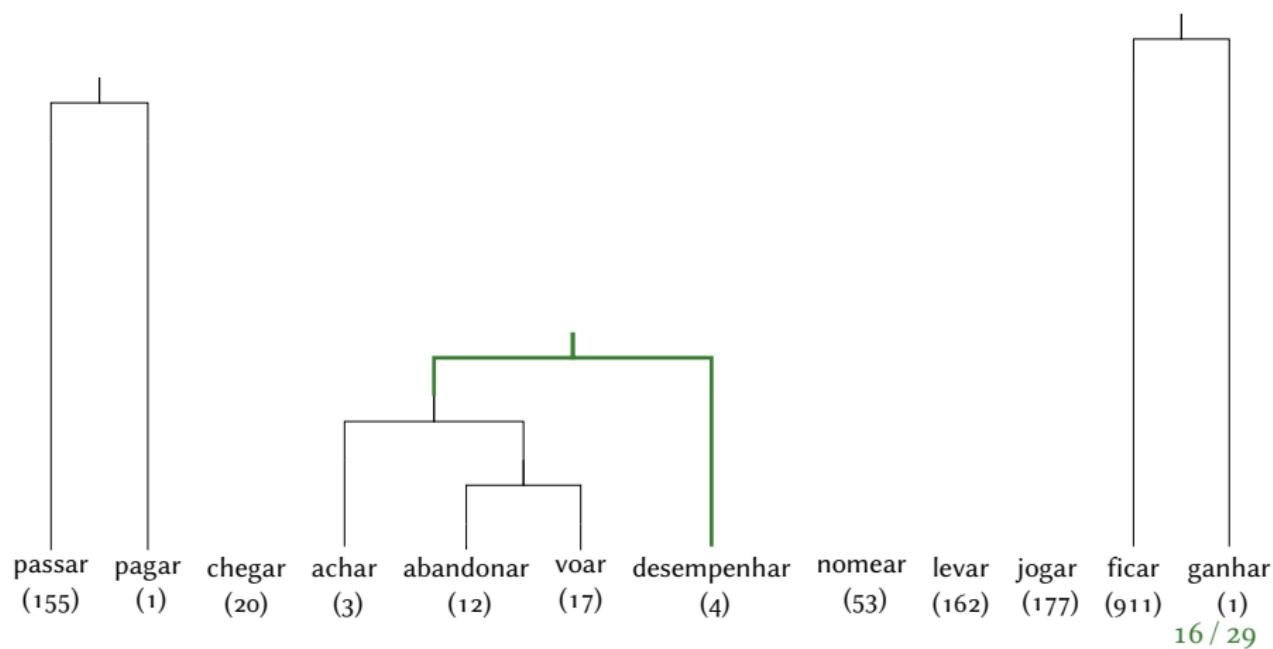
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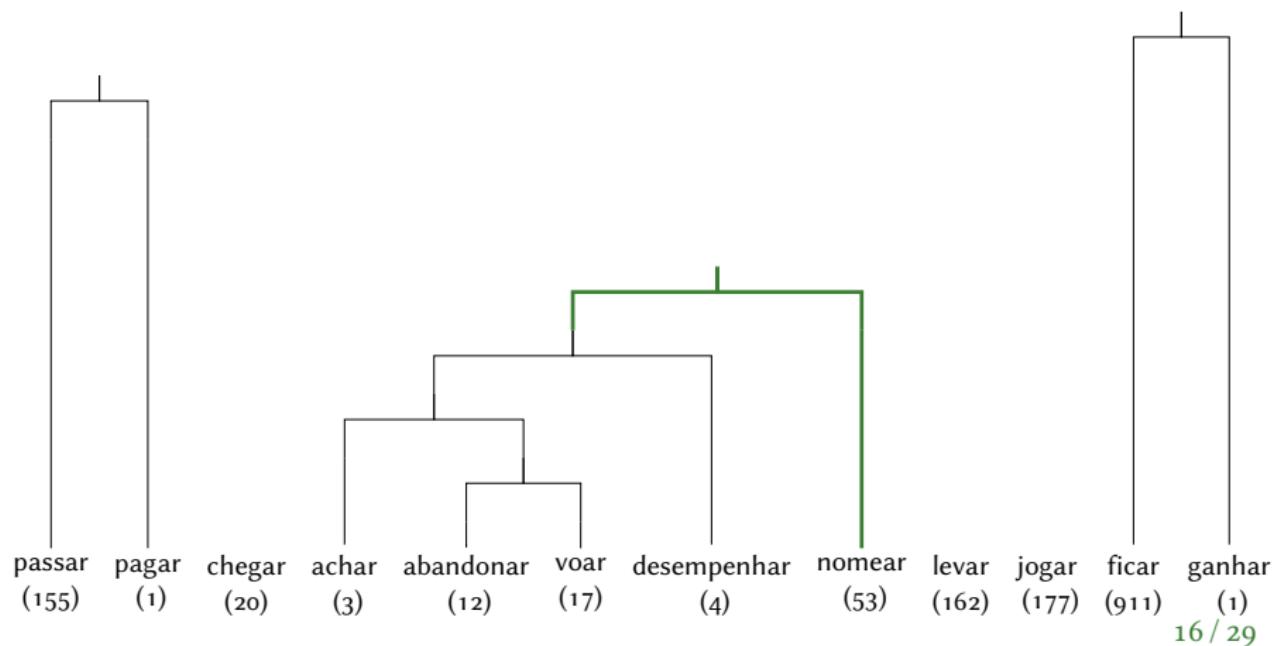
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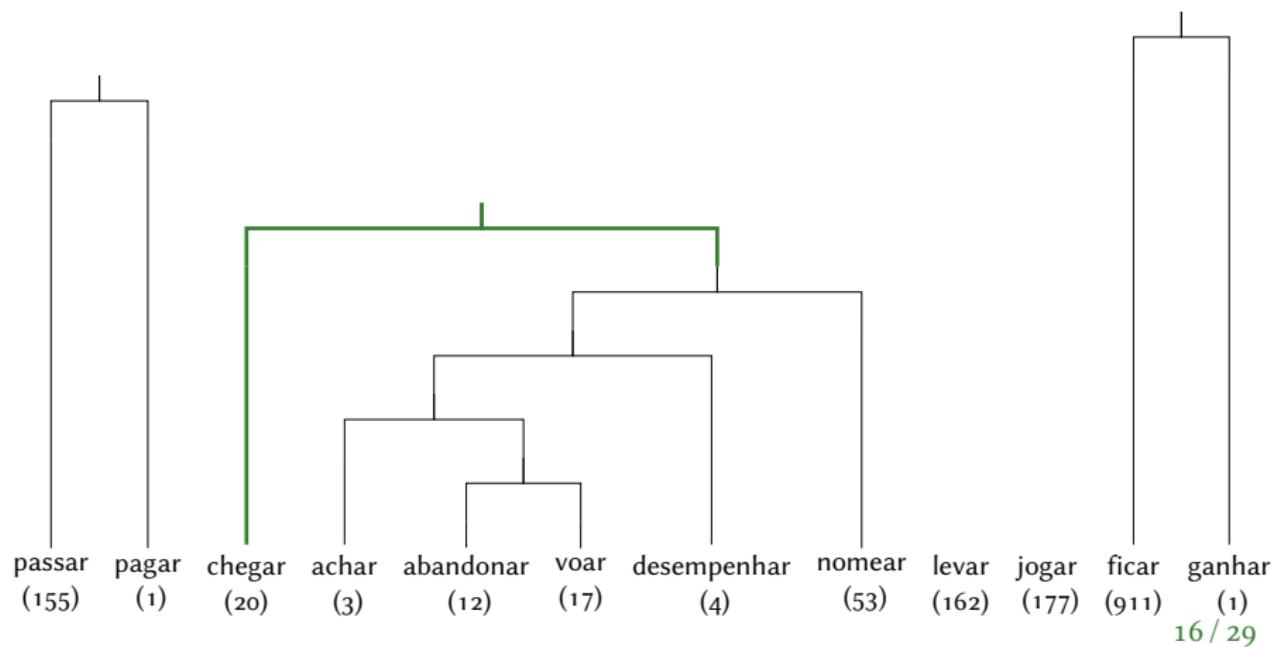
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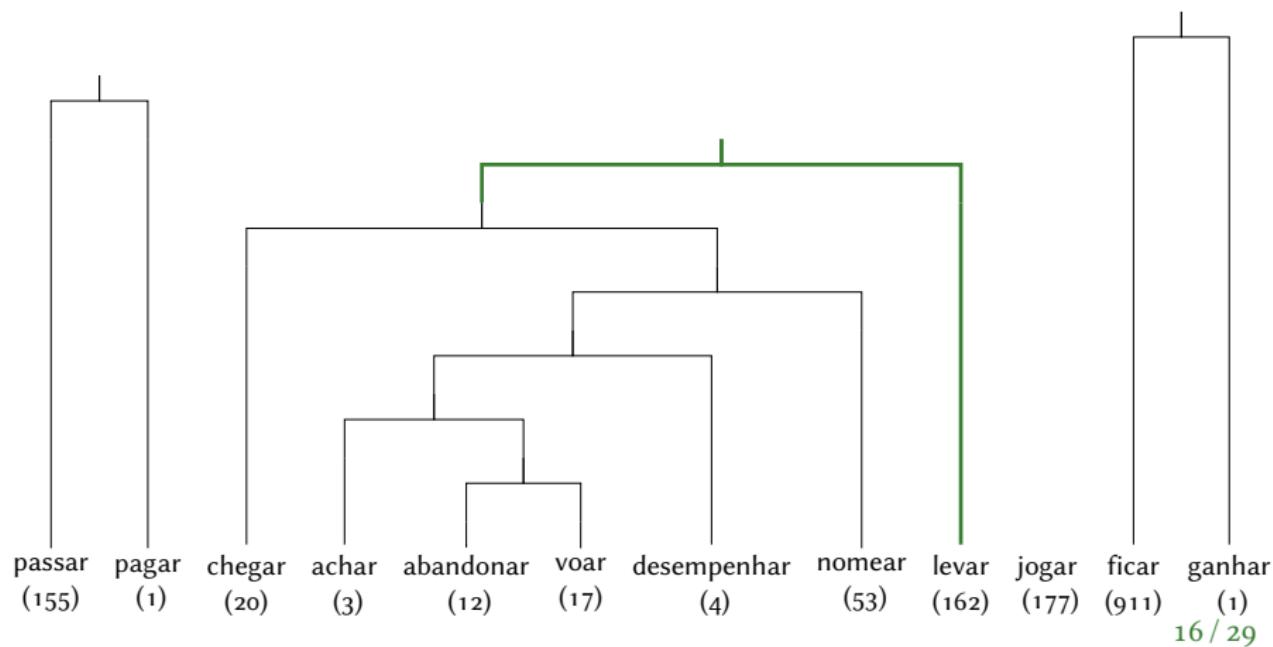
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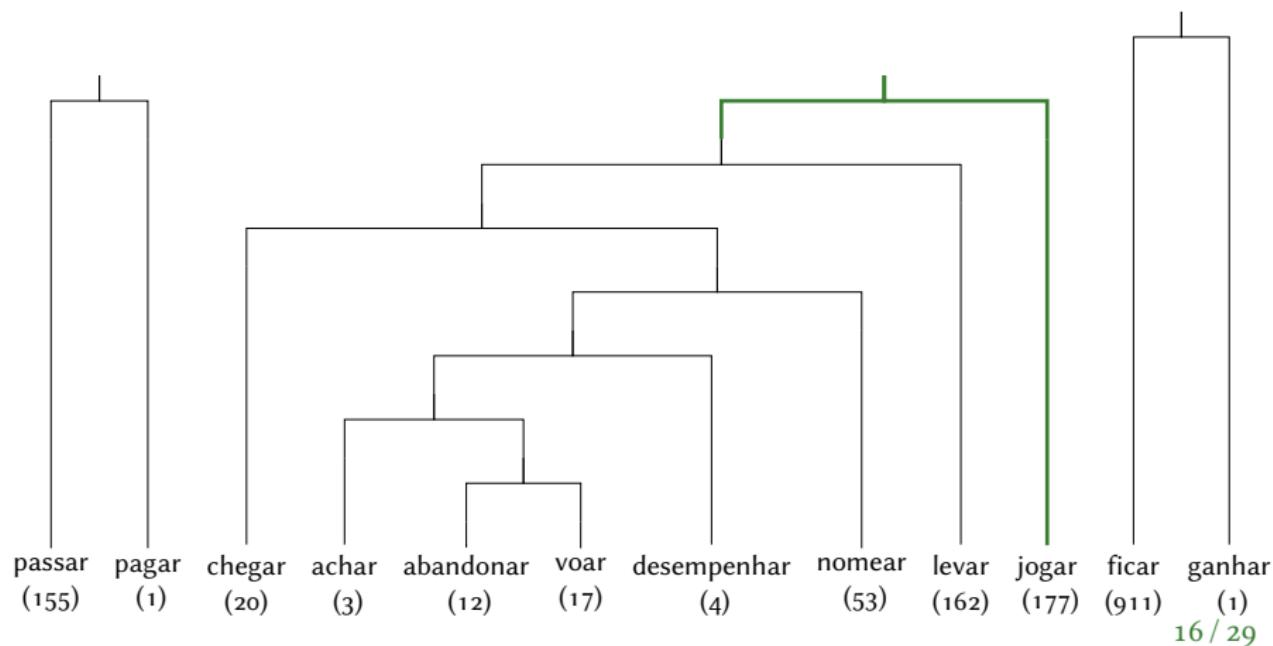
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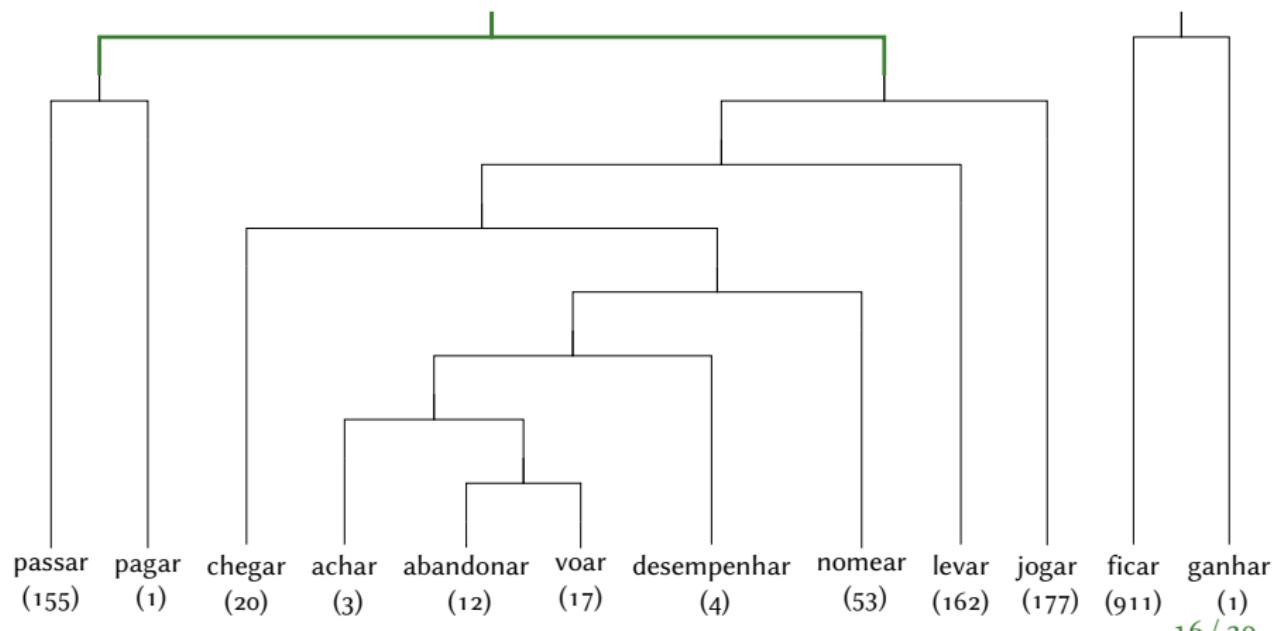
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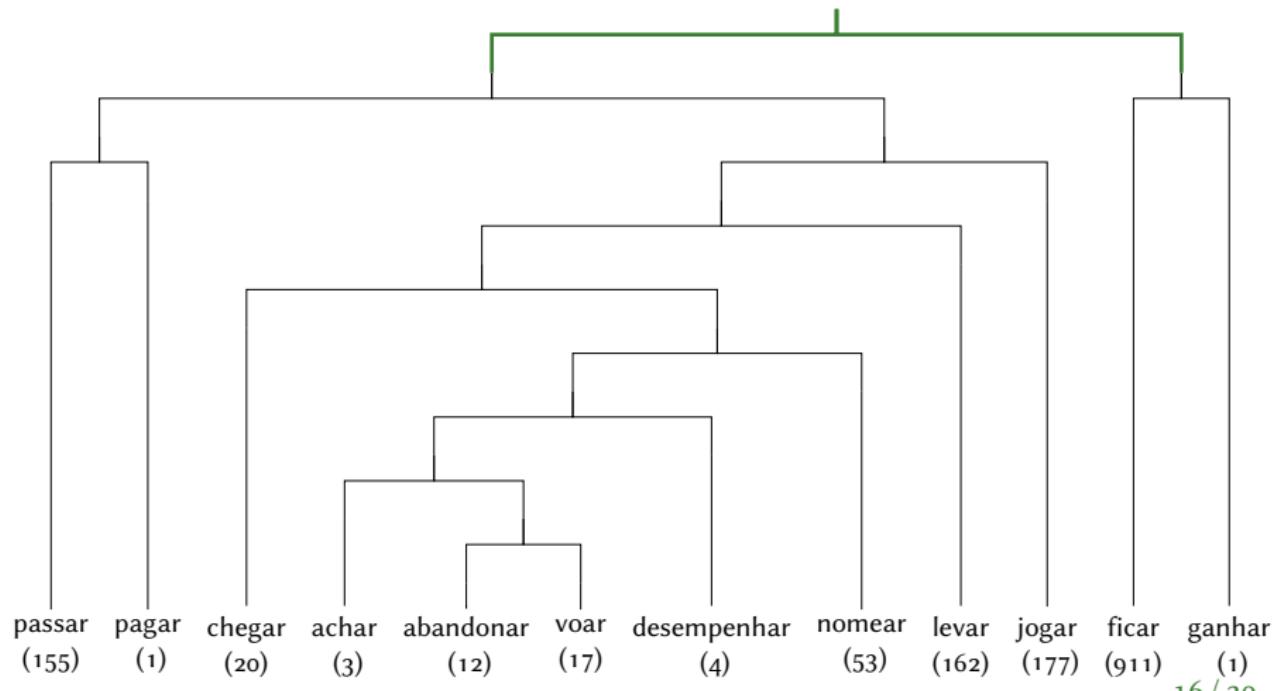
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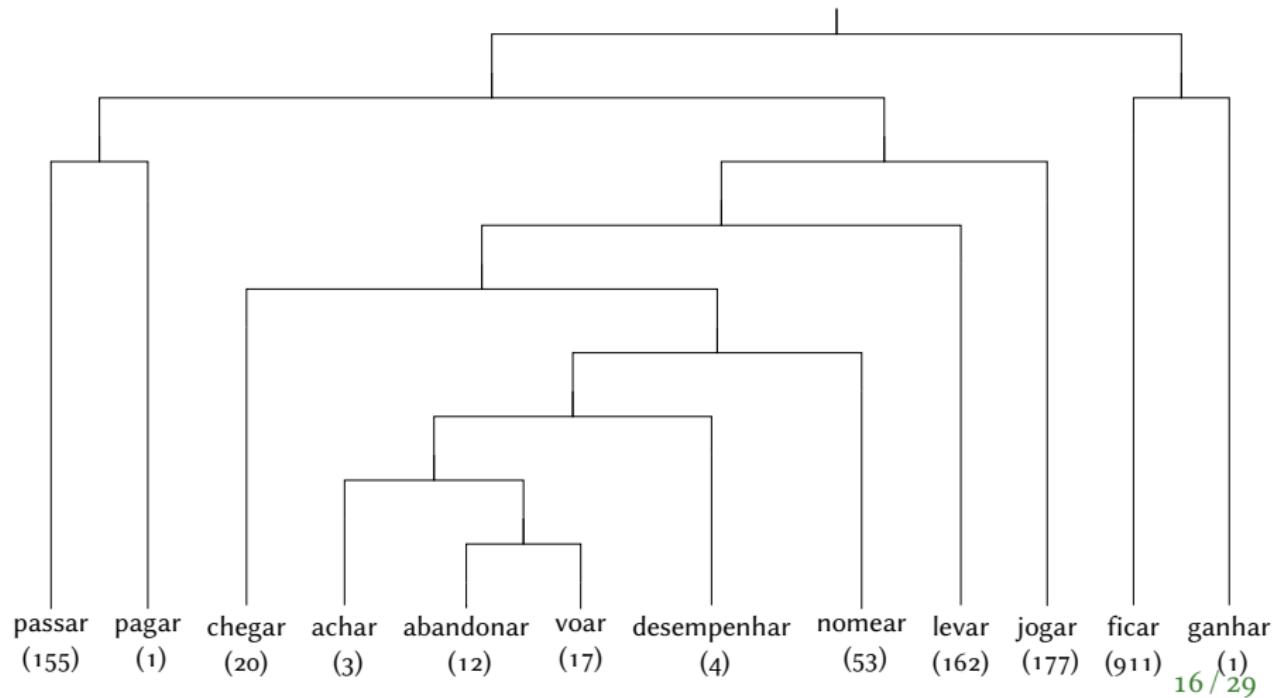
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HOW DO WE IDENTIFY MACROCLASSES ?

- ▶ As we merge clusters, we expect the DL to decrease.
- ▶ We define **macroclasses** when at some point, merging doesn't lead to a decrease.
 - ▶ Macroclasses optimise the size of the system, without any information loss.
 - ▶ It is an empirical issue whether a system has macroclasses or not.
- ▶ They do show up in our experiments.

COMPARISON TO OTHER WORKS

	Generalisations	Criterion	Algorithm
Brown and Evans (2012)	raw paradigms	Compression distance	CompLearn
Bonami (2014)	Affixes	Edit distance	UPGMA
Bonami (2014)	Patterns	Hamming distance	UPGMA
Lee and Goldsmith (2013)	Sets of characters	DL variant	greedy bottom-up
Here	Local patterns	DL	greedy bottom-up
Here	Global patterns	DL	greedy bottom-up

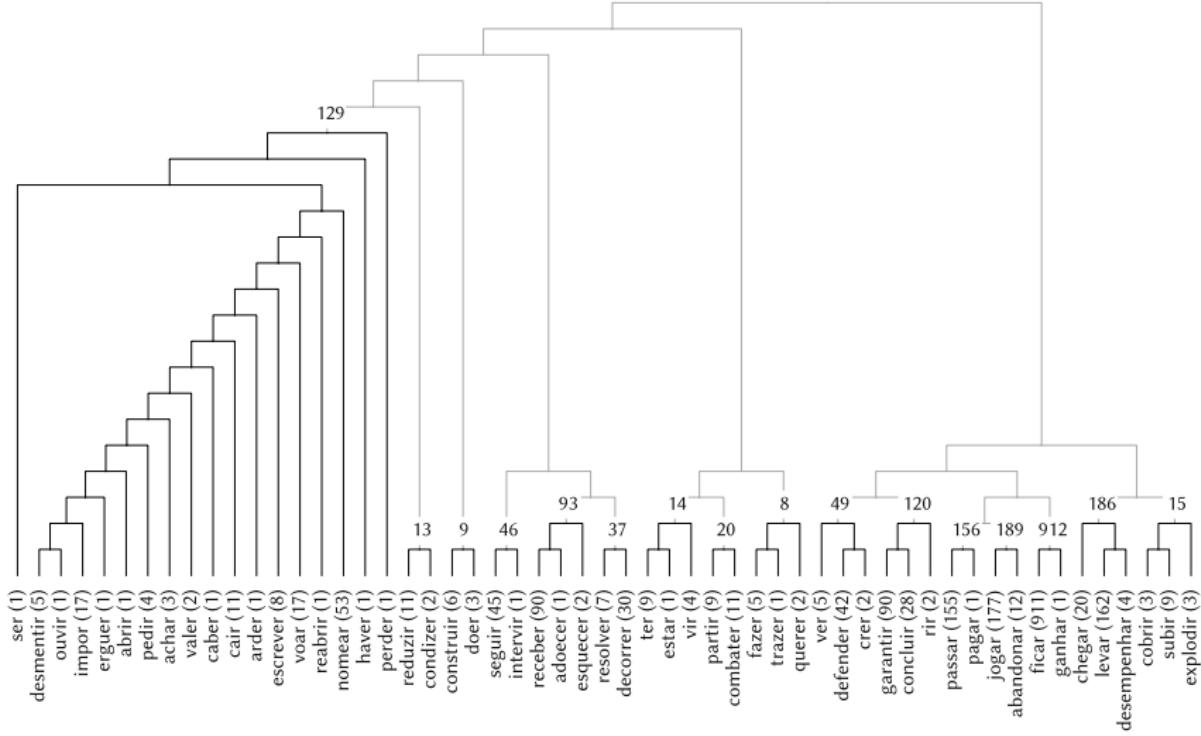
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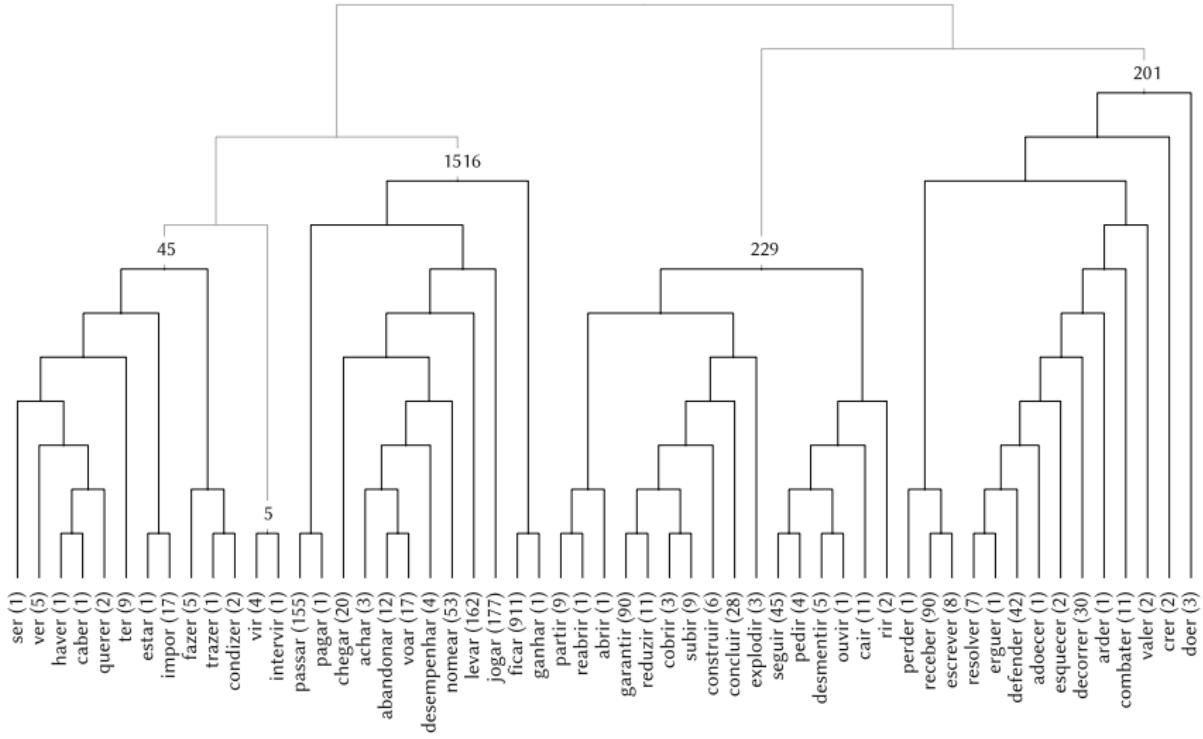
DATASETS

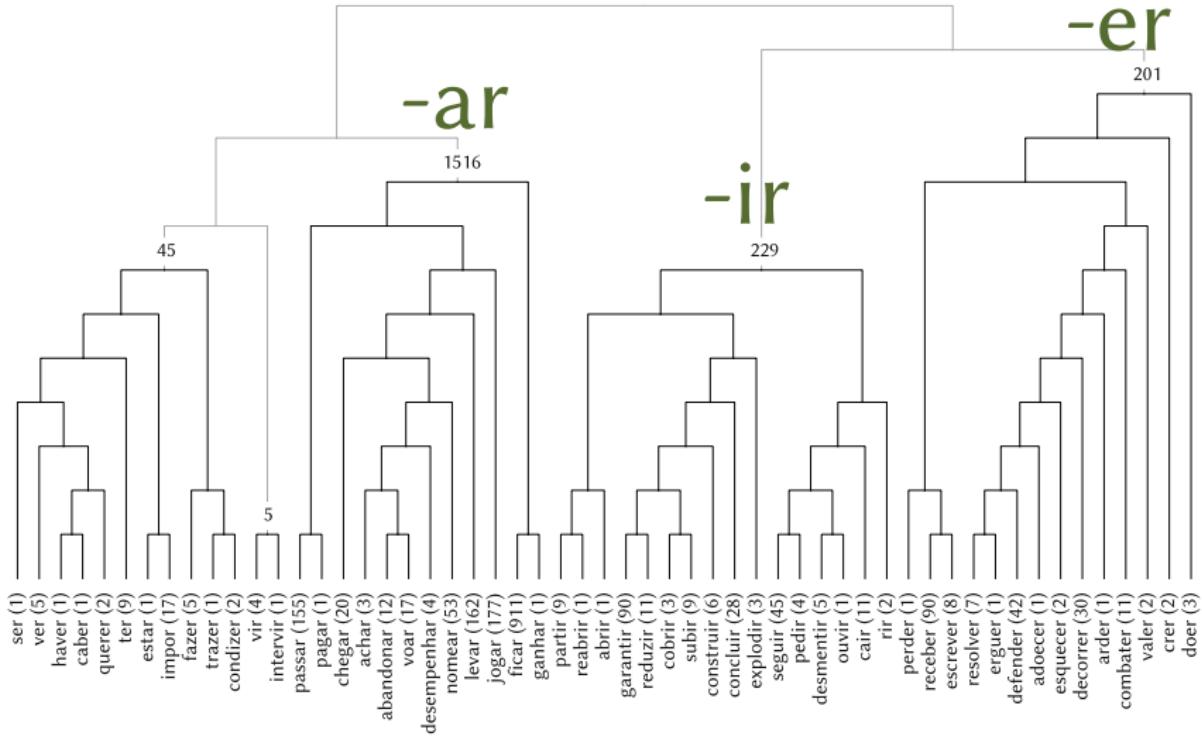
- ▶ Paradigm tables in phonemic transcription.
- ▶ **European Portuguese:** Coimbra pronunciation dictionary (Veiga et al., 2013) (1995 verbal entries).
- ▶ **French:** Flexique (Bonami et al., 2014) (5406 verbal entries).

Portuguese inflection classes, with global templates

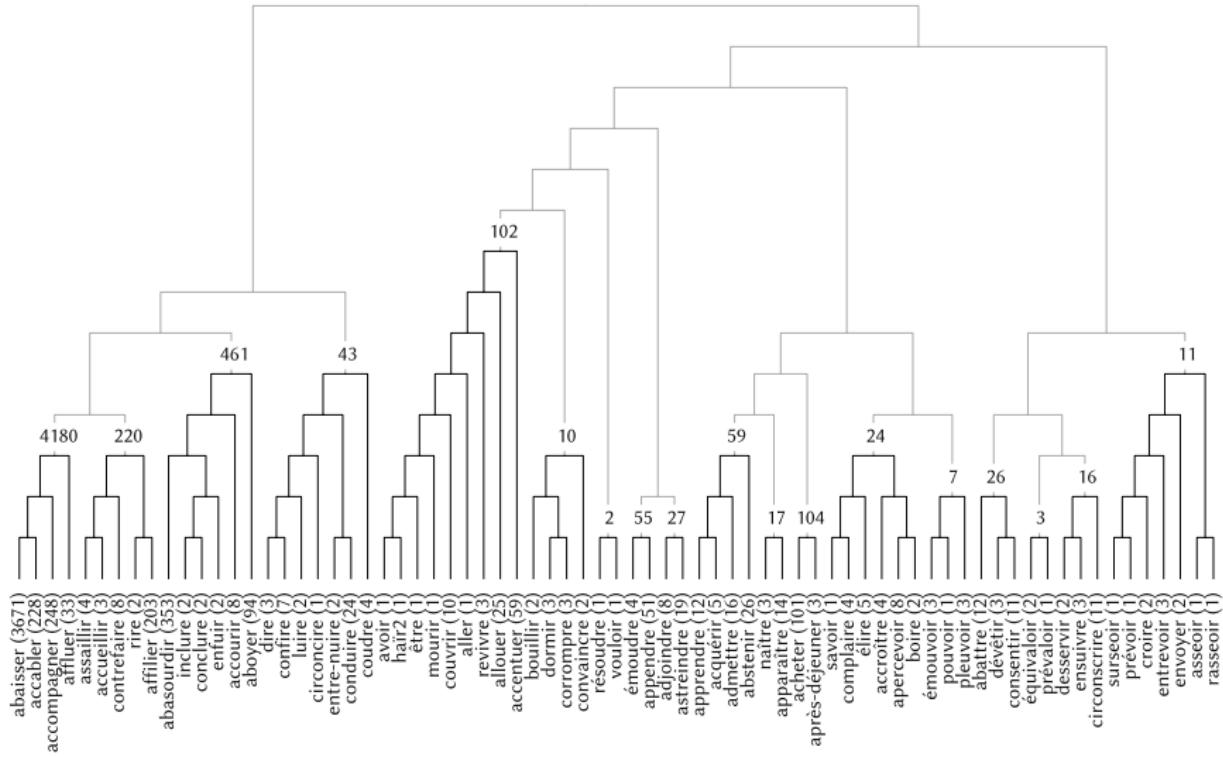


Portuguese inflection classes, with local templates





French inflection classes, with global templates



French inflection classes, with local templates

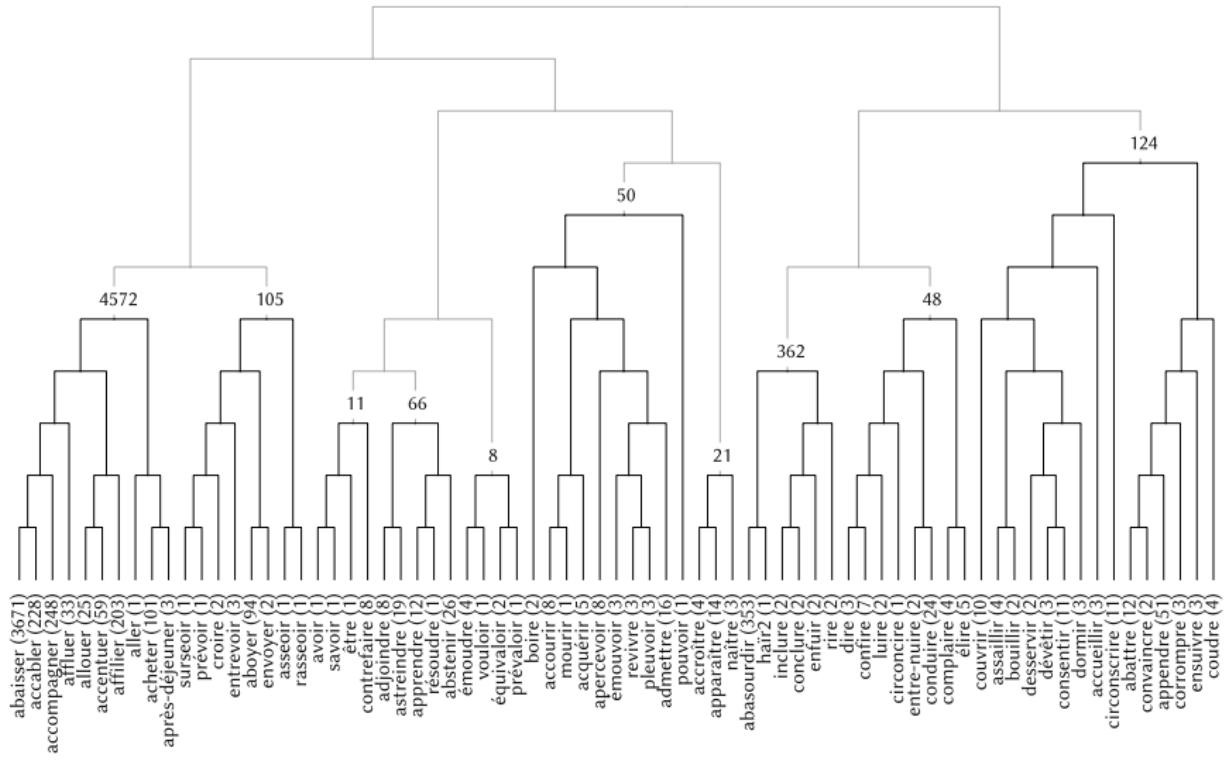


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CONCLUSION

- ▶ **Main properties :**
 - ▶ Based on information-theoretic measures.
 - ▶ Relies on automatically inferred generalisations.
- ▶ **Compared to previous attempts** to automatise IC :
 - ▶ Segmentations not taken as granted.
 - ▶ Using a measure that evaluates the quality of the system allows us to infer relevant macroscopic generalisations.
 - ▶ No parameters to adjust: **Occam's razor** is the only criterion.
- ▶ **Local segmentation** better captures the structure in inflection system than global segmentations.
 - ▶ Supports the relevance of local patterns of alternation in abstractive approaches (Blevins, 2006).
 - ▶ Complementary to work on information-theoretic modeling of implicative structure (Ackerman et al., 2009; Ackerman and Malouf, 2013; Bonami and Beniamine, 2015)
- ▶ **Freely available code** : <http://qumin.gforge.inria.fr>

Qumín

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